

How to Learn Math for Data Science, The Self-Starter Way

October 30, 2016

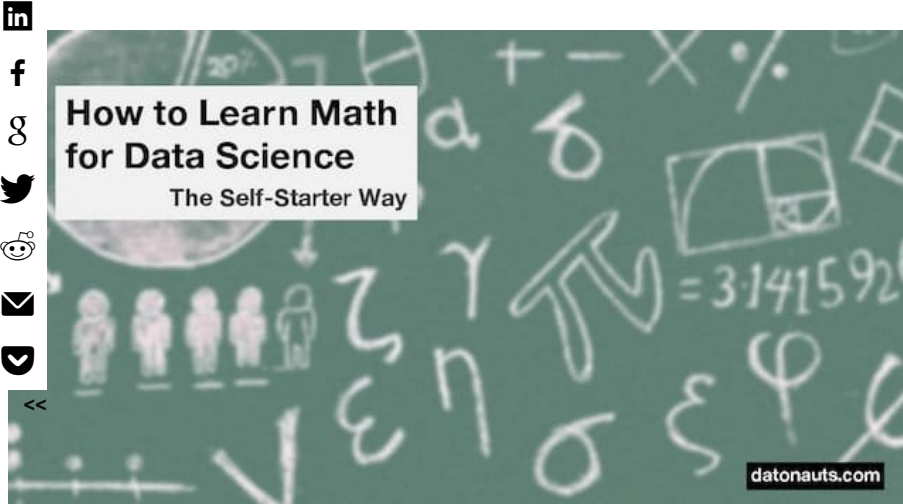
f Share

g Google

in LinkedIn

🐦 Tweet

Do you need to have a math Ph.D to become a data scientist? Absolutely not! This guide will show you how to learn math for data science and machine learning without taking slow, expensive courses.



How much math you'll do on a daily basis as a data scientist varies a lot depending on your role. Keep reading to find out which concepts you'll need to master to succeed for your goals.

Pre-requisite: Basic Python Skills

To complete this guide, you'll need at least basic Python* programming skills. We'll be learning math in an applied, hands-on way.

Check out our guide, [How to Learn Python for Data Science, The Self-Starter Way](#), for the fastest way to get up to speed with Python. We recommend at least completing up to **Step 2** in that guide.

**note: other languages are fine too, but the examples will be in Python.*

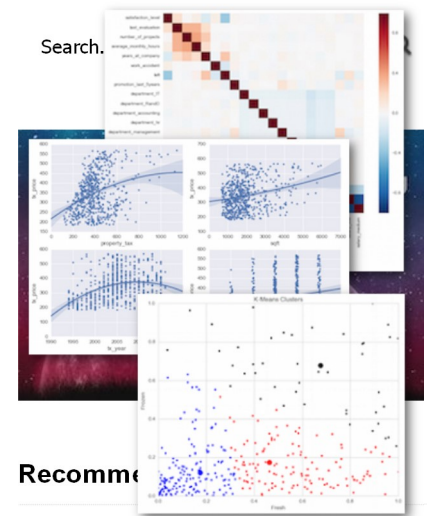
Math Needed for Data Science

The amount of math you'll need depends on the role. First, every data scientist needs to know some statistics and probability theory. We have a guide for that:

- [How to Learn Statistics for Data Science, The Self-Starter Way](#)

What about other types of math? Well, here's where the answer is more nuanced... it

How to Learn Practical Machine Learning in Just 1 Month...



Recommendations

[The 5 Levels of Machine Learning Iteration](#)
Skip the academics. Learn practical ML from [R vs Python for Data Science: Summary of Modern Advances](#)
[how to get real results with machine learning.](#)
[Python Machine Learning Tutorial, Scikit-Learn: Wine Snob Edition](#)
[Learn more about this 100% project-based Keras Tutorial: The Ultimate Beginner's Guide to Deep Learning in Python](#)

[21 Must-Know Machine Learning Interview Questions and Answers](#)

Supercharge Your Data Science Career!

Download Our Hand-Picked List of 88 Awesome Free Resources

GET INSTANT ACCESS

We'll never sell or spam your email.

depends on how much *original machine learning research* you'll be doing.

How to Learn Practical Machine Learning in Just 1 Month

In practice, especially in entry-level roles, you'll often be using out-of-the-box ML libraries of common libraries in many programming languages. You'll be able to peel back ML algorithms and work with the wheel.

Even so, you'll need to build custom implementations of ML algorithms. You may need to adapt one to your tech stack or to expand its capabilities. You'll need to be able to peel back ML algorithms and work with the wheel.

When you're in a position where you need to build custom implementations of ML algorithms, you'll need to adapt one to your tech stack or to expand its capabilities. You'll need to be able to peel back ML algorithms and work with the wheel.

LinkedIn
Facebook
Google+
Twitter
Reddit
Email
Medium

For entry-level roles, you'll often be using out-of-the-box ML libraries of common libraries in many programming languages. You'll be able to peel back ML algorithms and work with the wheel.

For these positions, **mastery of both linear algebra and multivariable calculus** is a must. You'll need to be able to peel back ML algorithms and work with the wheel.

The Best Way to Learn Math for Data Science

The self-starter way to learning math for data science is to **learn by "doing shit."** So we're going to tackle linear algebra and calculus by using them in real algorithms!

Even so, you'll want to learn or review the underlying theory up front. You don't need to read a whole textbook, but you'll want to learn the key concepts first.

Here are the 3 steps to learning the math required for data science and machine learning:

- 1 **Linear Algebra for Data Science**
Matrix algebra and eigenvalues.
- 2 **Calculus for Data Science**
Derivatives and gradients.
- 3 **Gradient Descent from Scratch**
Implement a simple neural network from scratch.

Step 1: Linear Algebra for Data Science

Many machine learning concepts are tied to linear algebra. For example, **PCA** requires eigenvalues and **regression** requires matrix multiplication.

Also, most ML applications deal with high dimensional data (data with many variables). This type of data is best represented by matrices.

Here are a few of the best free resources we've found for learning linear algebra for data

science:



For application-heavy roles...

[Khan Academy](#) has short, practical linear algebra lessons. They cover the most important topics.

For R&D-heavy roles...

[MIT OpenCourseWare](#) offers a rigorous linear algebra class. The video lectures and course materials are all included.

For the academics. Learn practical ML from practicing professionals who will teach you

If you only need to review:

How to get real results with machine learning. [Linear Algebra Review for Machine Learning \(Video Series\)](#) - These are the optional

linear algebra review videos for Andrew Ng's machine learning course. The entire series can be watched in under 1 hour. Recommended if you've taken linear algebra before and just need a quick review.

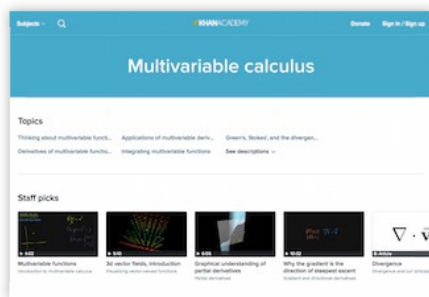
- [The Matrix Cookbook \(PDF\)](#) - Excellent reference resource for matrix algebra.

Step 2: Calculus for Data Science

Calculus is important for several key ML applications. For example, you'll need to be able to calculate derivatives and gradients for optimization.

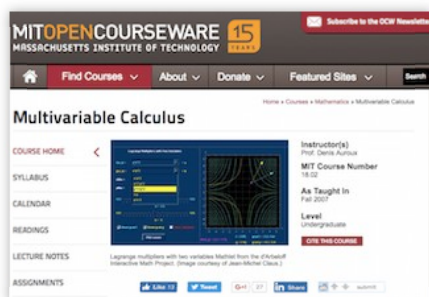
In fact, one of the most common optimization techniques is [gradient descent](#).

Here are some of the best resources for learning calculus for data science:



For application-heavy roles...

[Khan Academy](#) has short, practical multivariable calculus lessons. They cover the most important concepts.



For R&D-heavy roles...

[MIT OpenCourseWare](#) offers a rigorous multivariable calculus class. The video lectures and course materials are all included.

And if you only need to review:

How to Learn Practical Machine

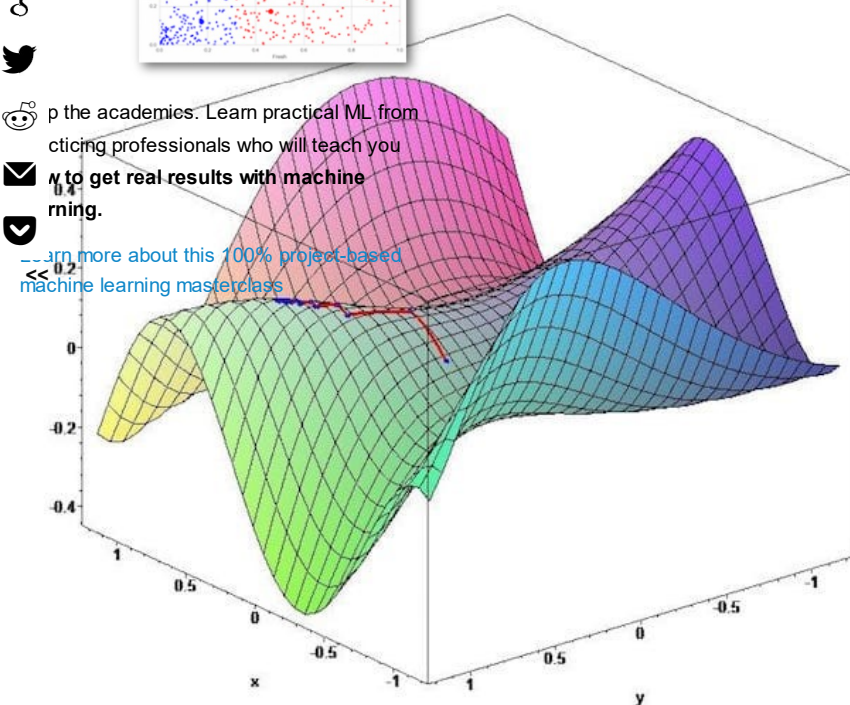
[Learning in Just 1 Month...](#) [Multivariable Calculus Review \(Video\)](#) - This is quick review of multivariable calculus in the format of solving practice problems. Recommended if you've taken multivariable calculus before and just need a quick review.

Step 1: Building a Neural Network from Scratch

Out of the way. Now it's time for the really fun part. The first step in learning data science and machine learning is to build a neural network. We'll use linear algebra, calculus, and the network and calculus to optimize it. Specifically, we'll use gradient descent from scratch.

Join the academics. Learn practical ML from practicing professionals who will teach you how to get real results with machine learning.

[Learn more about this 100% project-based machine learning masterclass](#)



Don't worry too much about the nuances of neural networks for now. It's ok if you're just following instructions and writing code. We'll cover machine learning in depth in another guide, as this is for targeted math practice.

Follow along with the tutorials, and review theory as you go along. Plus, you'll have a cool project to add to your portfolio afterward.

Here are a few awesome step-by-step guides:

- [Neural Network in Python, Part 2](#) - This is an incredible tutorial that takes you through a simple neural network from end to end. It's packed with helpful illustrations, and you'll learn about how gradient descent fits in.
- [Neural Nets to Recognize Handwritten Digits](#) - We love this resource! This is a free online book that walks you through a famous application of neural networks. It explains ideas very intuitively, and it's the most in-depth tutorial in this list.
- [Implementing a Neural Network from Scratch](#) - A shorter tutorial that also takes you

through step-by-step.

How to Learn Practical Machine Learning in Just 1 Month

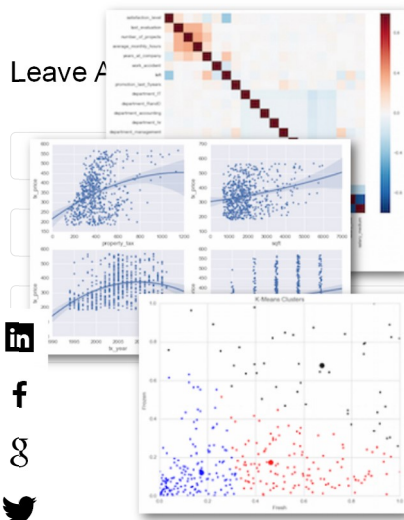
Share

Google

LinkedIn

Tweet

Leave A



Name*

Email*

Website



Join the academics. Learn practical ML from practicing professionals who will teach you

How to get real results with machine

Learning.
Submit Comment

* Denotes Required Field

[Learn more about this 100% project-based machine learning masterclass](#)

Copyright © 2016 · EliteDataScience.com · All Rights Reserved

[Home](#) [Terms of Service](#) [Privacy Policy](#)